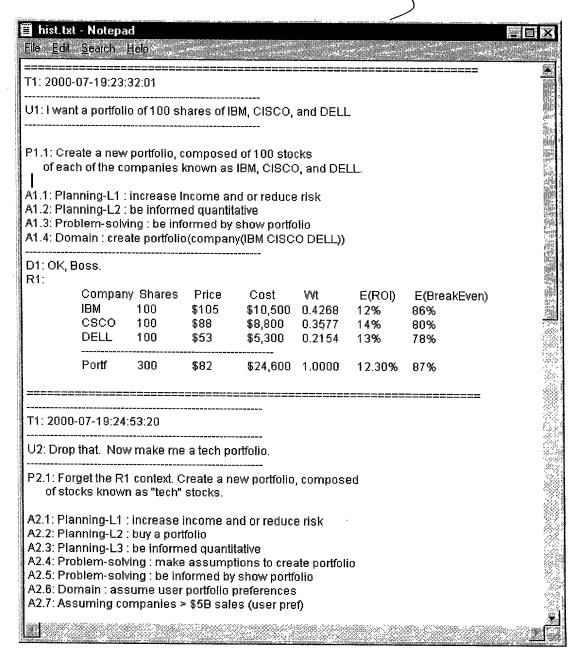


1302

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1301 1303 賽IPE V 0.683 - Netscape File Edit View Go Communicator Help Dialog: Visible Thought: A3.1: Planning-L1: increase income and or reduce risk
A3.2: Planning-L2: buy a portfolio
A3.3: Planning-L3: be informed quantitative
A3.4: EM-Discourse: cursor on 2000-07-19.R2 indicates "that
A3.5: UserPreferences . support user dislakes ED3, p = 0 55 UL: I want a portfolio of 100 shares of IBM, CISCO, and DELL U2: Drop that. Now make me a terh portfolio U3: Try it without that (gestures at 2000-07-19.R2.Company=EDS)
U4: What is the misk profile P4 L: Display a graph of a simulation of financial outcomes A4 1: Plarming-L1: increase intome and or reduce risk
A4.2: Plarming-L2: buy a portfolio
A4.3: Plarming-L3: be informed quantitative
A4.4: Discourse: "that" is reference to 2000-07-19.R3
A4.5: Problem-solving: assume default simulation parameters
A4.5: Domain: create, graph outcome simulation(2000-07-19 R3) Results: D4: Here it is. Simulation: Portfolio v. Time (Quarters) \$45,000 \$40,000 Ŗ4: \$35,000 \$30,000 \$25,000 \$20,000 0 10 15 ET =0-1

1401



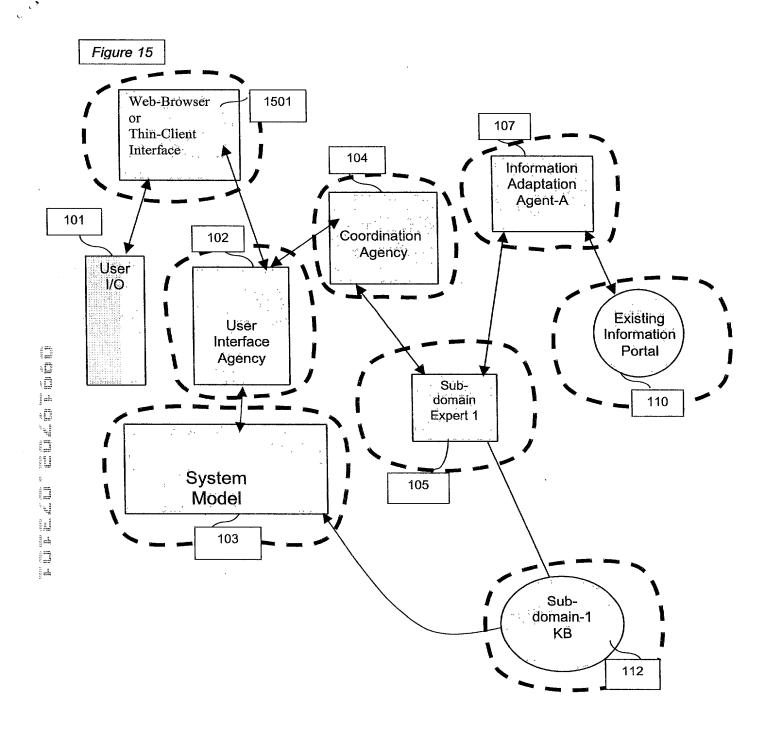


Figure 16

Simplified Strength/Necessity Belief Calculus

?X = Coffee if:

?X is in a mug (s = .2; n = 0)?X is a hot liquid (s = .4; n = 0)?X is brown (s = .6; n = .97)?X is not tea (s = .3; n = 1)

S = Strength; N = Necessity; B = Belief; D = Disbelief;

P = Belief measure of premise (input)

Belief Evaluation Recurrence Formulae:

$$\begin{array}{ll} B_{x+1} = B_x + (1 - B_x) * S_{x+1} * P_{x+1} & ; \text{ with } B_o = 0 \\ D_{x+1} = D_x + (1 - D_x) * N_{x+1} * (1 - P_{x+1}) & ; \text{ with } D_o = 0 \end{array}$$

$$\begin{array}{ll} \text{Conclusion} = B_n * (1 - D_n); \end{array}$$

Example A. $B_4 = 0.8656$, given all 4 preconditions known to be true with absolute certainty.

Example B.
$$B_4 = 0.7648$$
, $D_4 = 0.485$, Conclusion = 0.393872,

given that we are only 50% sure that the liquid is brown, but are convinced of all other facts (e.g. because the light is very dim....)

Bayesian Belief Calculus -

Bayes's rule states that:

$$p(A | B) = Prob \text{ of event A, given event B}$$

= $(p(A) * p(B | A)) / p(B)$

If we know the probabilities B_i for *every* way that A may be realized, we may write:

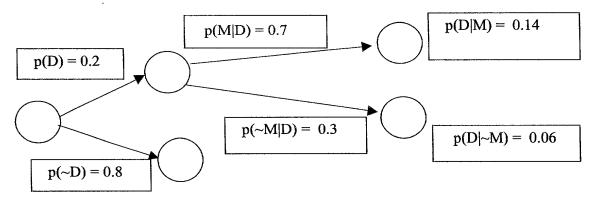
$$p(A) = \sum p(A|B_i) p(B_i)$$

Which allows a straightforward way to compute likelihood, when all possibilities are accounted for.

We can construct networks which relate Bayesian likelihood to various conditions. For example, consider the case where we are given

$$p(D)$$
 = probability of planning for retirement = 0.2, and $p(M \mid D)$ = probability of asking about mutual Funds, given D, = 0.7.

Now we can construct a graph of probabilistic influences that can be inferred:



This mechanism can be used to connect the probabilities of various plans and alternatives, and to infer likely plans from various communications.

Figure 18

